AI for Urban Climate: Using **EO-Based and Community data** for Air temperature downscaling at Urban Scales

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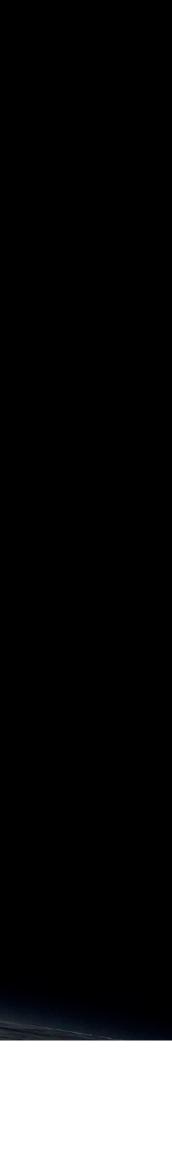
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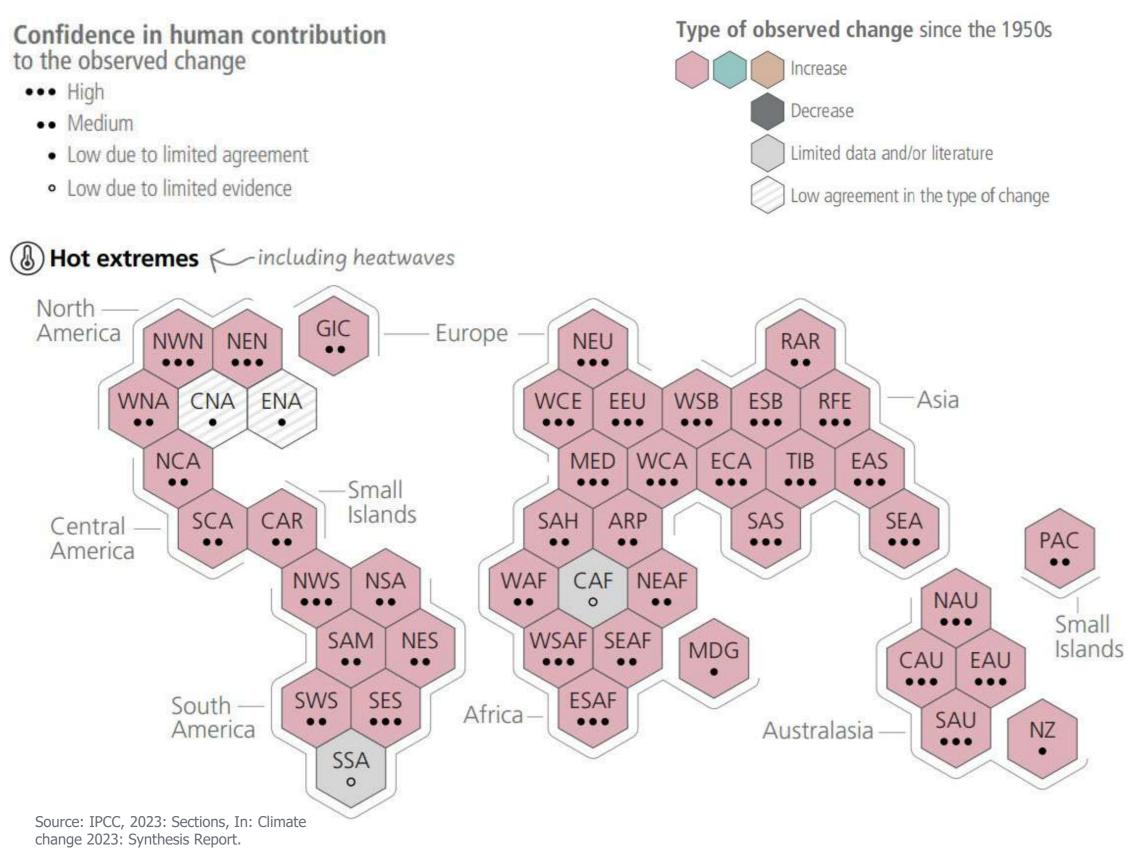


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Extreme Heat Context and Risks





Source: European Climate and Health Observatory (www.climate-adapt.eea.europa.eu) Accessed at: 2024/05/27



Over 4% of Summer Mortality in European Cities is Attributable to Urban Heat Islands

Source: Barcelona Institute for Global Health (www.isglobal.org) Accessed at: 2024/05/27





Motivation

Important questions





How much cooler/warmer is a neighbourhood, compared to the climate average?



How extreme is the heat/cold in each neighbourhood?





Which are the cooling/heating acclimatization needs in each neighbourhood?



Preliminary considerations

AROME ML Downscaling model

Existing Limitations

- **AROME** downscales global model to **regional scale**
- WMO stations are located to have minimum urban contributions
- Netatmo stations have variability associated to citizen owned data and a known positive bias

Goals



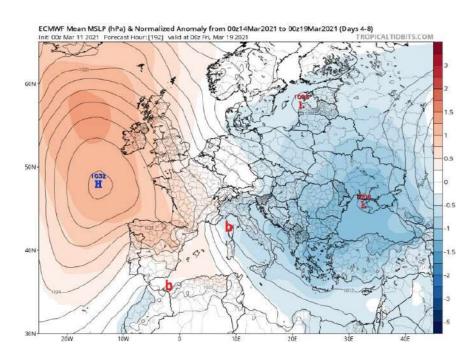
- **Downscale AROME** by 10x (from 2.5 Km to 250 m)
- Maintain or Improve the **AROME native error/bias**
- Depict physically known thermal response (urban vs. green infrastructure)
- Uphold or improve AROME deviations during **heat/cold** extremes

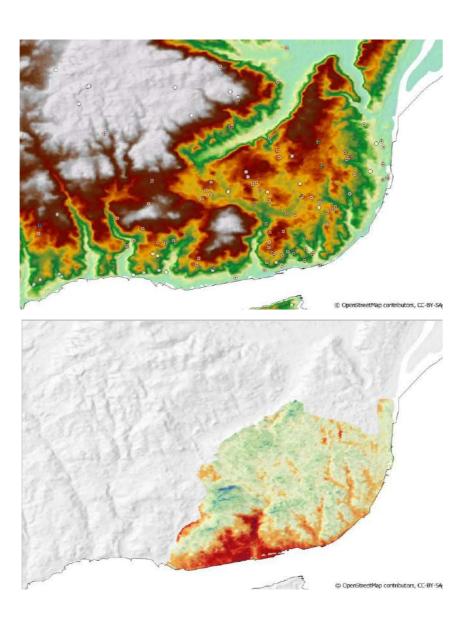


Preliminary considerations

Model Formulation

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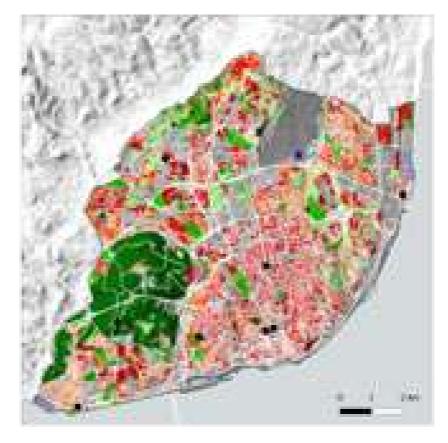


Background Conditions

Air Temperature Mean Sea Level Pressure Relative Humidity Wind U/V

Landscape Effects

Elevation TOPEX (per wind U/V) Coast/river distance ╋







Urban Effects

Bowen Ratio (QH/QE) Imperviousness Tree-cover percentage

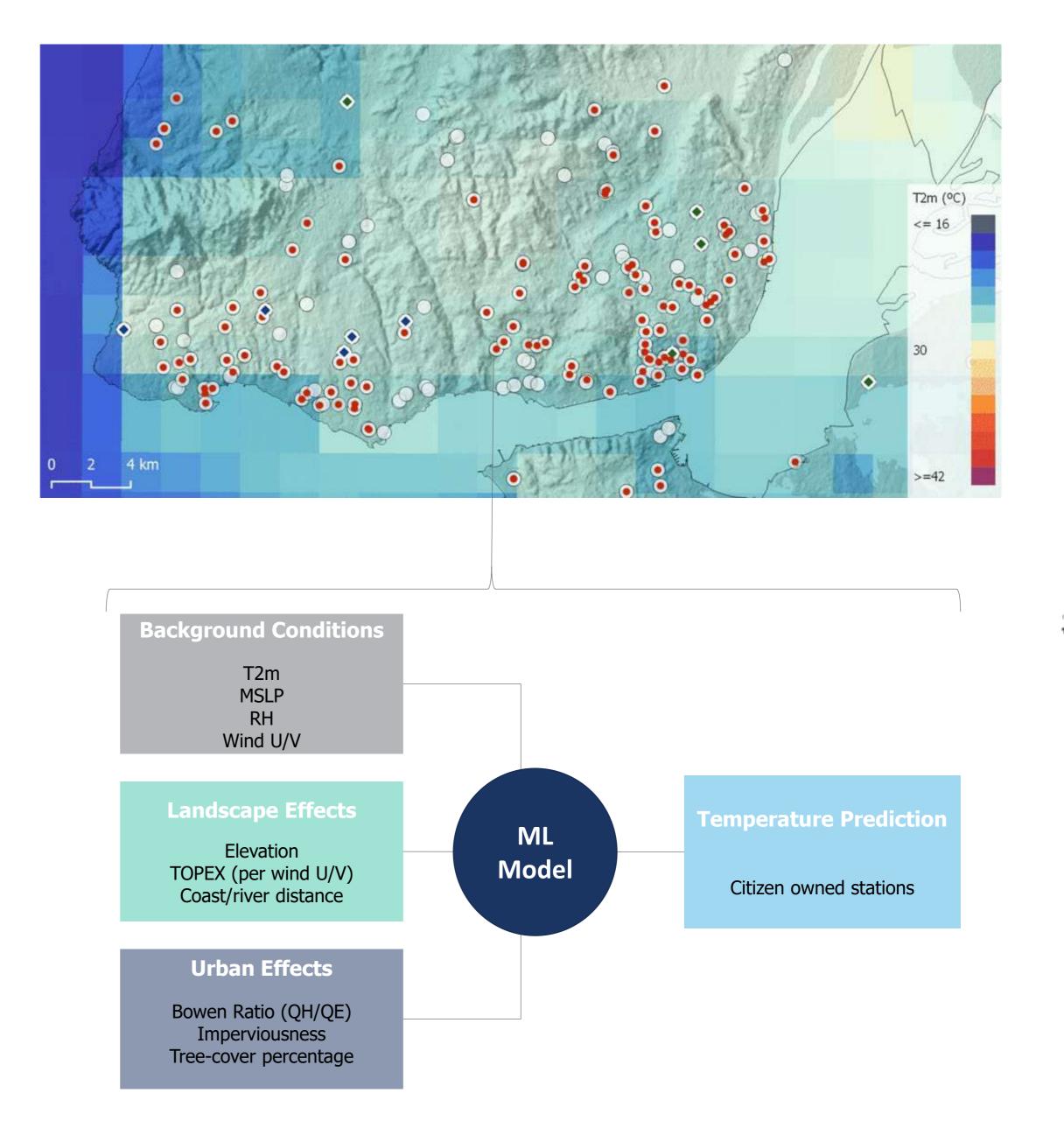
Temperature Prediction

Citizen owned stations (Quality Controlled)



Preliminary Considerations

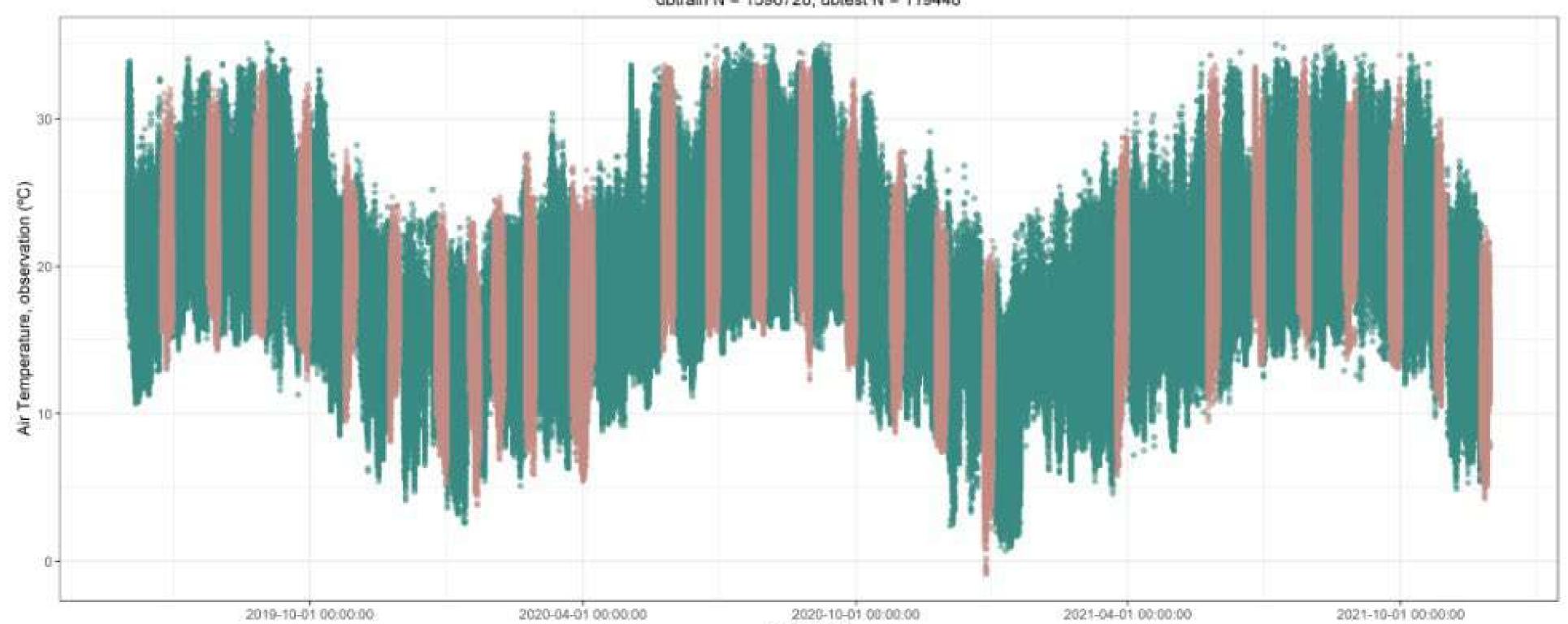
Model Formulation







Netatmo data Time Series





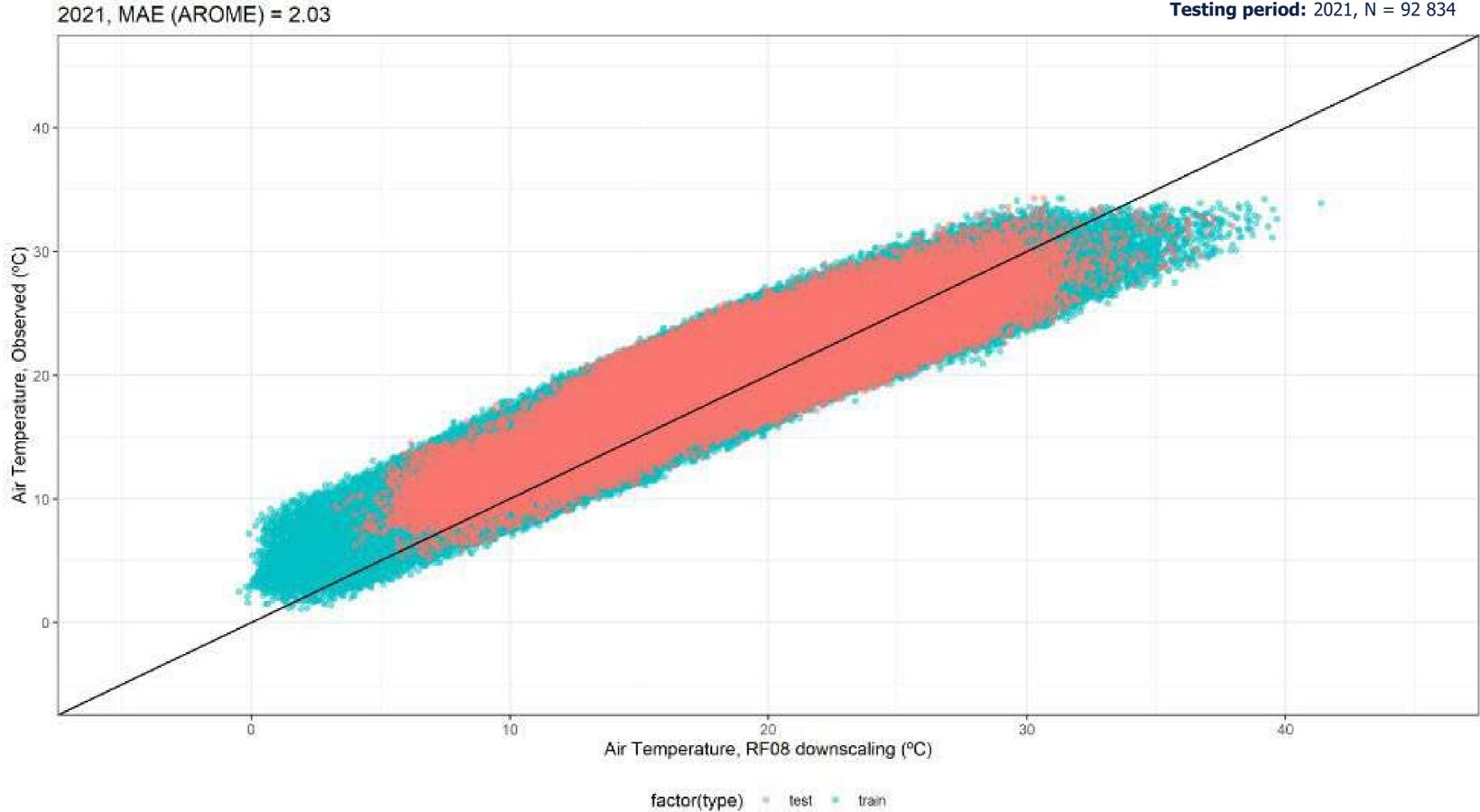
Time series of Netatmo input data - after quality control dbtrain N = 1596726, dbtest N = 119448



Date time



Netatmo temperature error



Training period: 2021, N = 506 951 **Testing period:** 2021, N = 92 834





Results

Performance RF vs AROME

Training period: 2021, N = 506 951 **Testing period:** 2021, N = 92 834

Month (2021)	MAE RF08, train (°C) (N = 506 951)	MAE RF08, test (°C) (N = 92 834)	MAE AROME (°C) (N=599 785)
DJF	1.32	-	1.74
MAM	1.75	1.80	2.12
JJA	1.75	1.67	2.18
SON	1.70	1.55	1.98

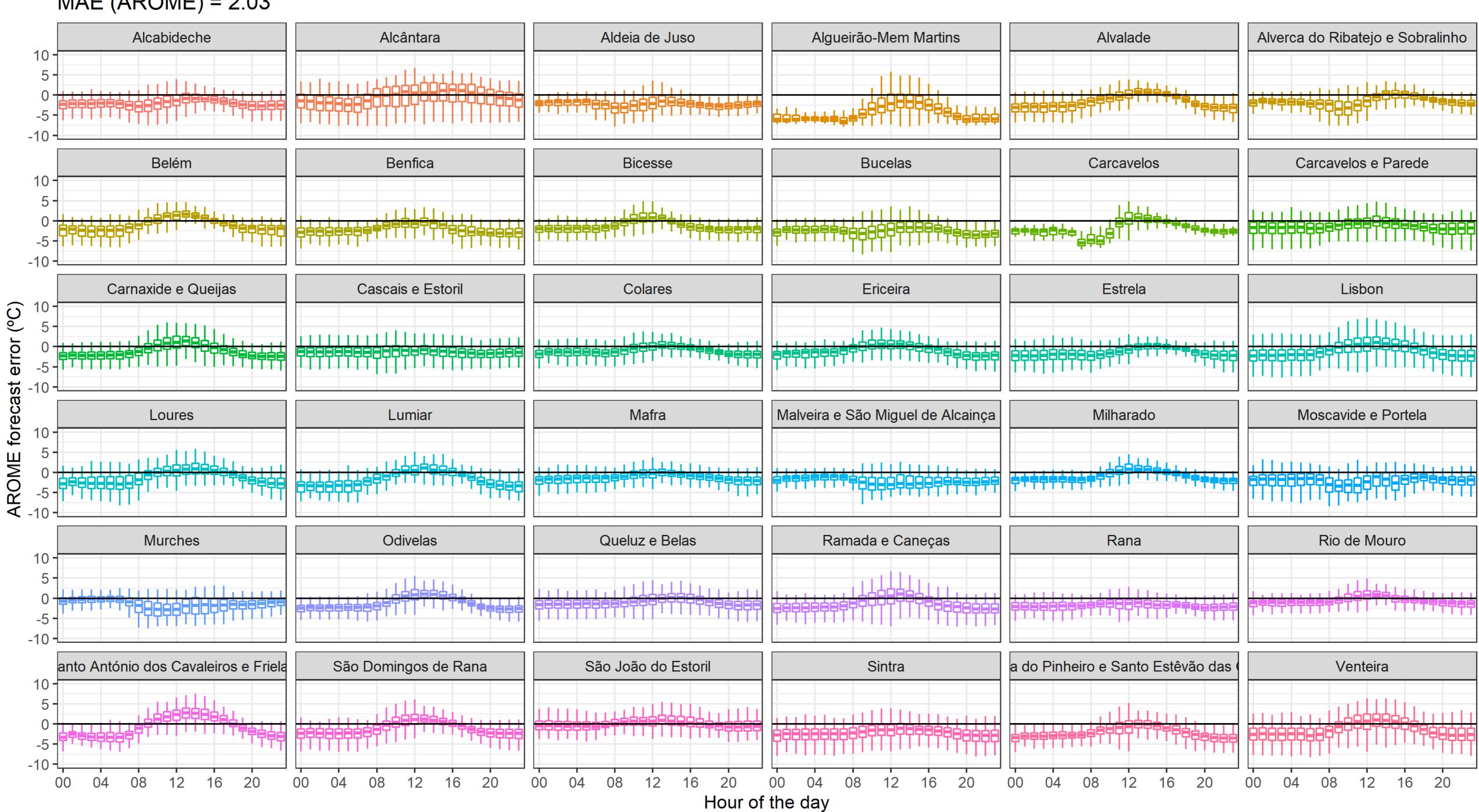


~ 0.4°C AROME bias reduction, ~20% accuracy improvement

Month (2021)	MAE RF08, train (°C) (N = 506 951)	MAE RF08, test (°C) (N = 92 834)	MAE AROME (°C) (N=599 785)
1	1.29	-	1.80
2	1.36	-	1.68
3	1.64	1.70	2.12
4	1.87	-	2.14
5	1.67	1.91	2.11
6	1.83	1.79	2.19
7	1.75	1.52	2.15
8	1.67	1.78	2.21
9	1.54	1.72	2.11
10	1.72	1.53	1.88
11	1.87	1.34	1.92
12	-	-	-







MAE (AROME) = 2.03

MAE (RF08) = 1.65 Alcabideche Aldeia de Juso Alcântara 10 5 0 **◆◆**◆◆◆◆◆ **╡╡╡╡╡╡╡╡**╡**┥** -5 --10 Belém Benfica **Bicesse** 10 5 ~~~~~ 0 **┩╡╡┽┽╷╷╷ ġġġġġġġġġġġ**ŢŢŢŢŢŢ<u>Ċ</u>ĊĊġġġġġ **╞┿**╪┿╪┿╪┿[╋] -5 -10 Carnaxide e Queijas Cascais e Estoril Colares RF08 downscaling error (°C) 10 5 **⋵**⋪⋵⋪⋵⋵⋵⋵⋳⋳⋳⋳⋴∊⋎⋎⋎⋳⋳⋴⋪⋪∊ ╹┙╾┑╴┰┼┾┾┾┾┿╡ -5 -Mafra Loures Lumiar 10 5 **╡**╪╡╡╡╡╡╡[╡]┥╵╵╵╵╵╵╴╸╸╡ -5 Murches Odivelas Queluz e Belas 10 5 0 -5 -' -10 anto António dos Cavaleiros e Friela São Domingos de Rana São João do Estoril 10 -5 -0 -5 --10 -4

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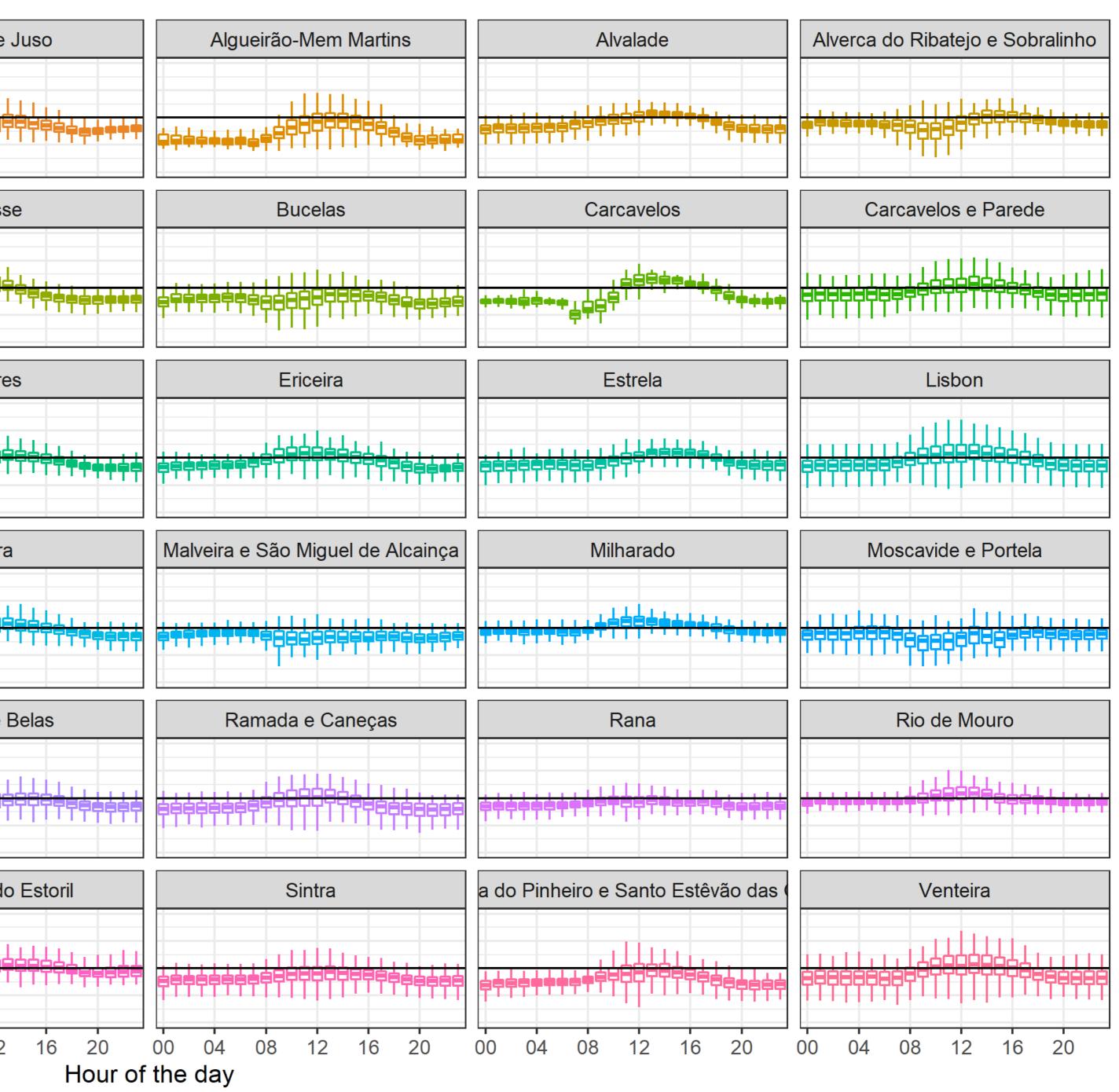
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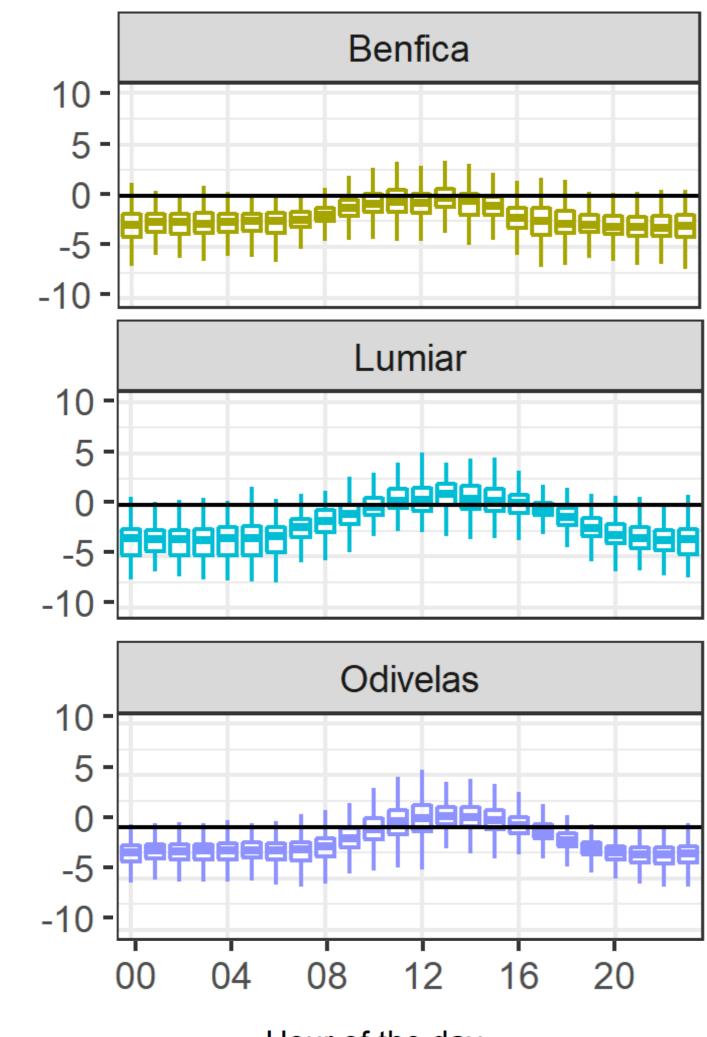
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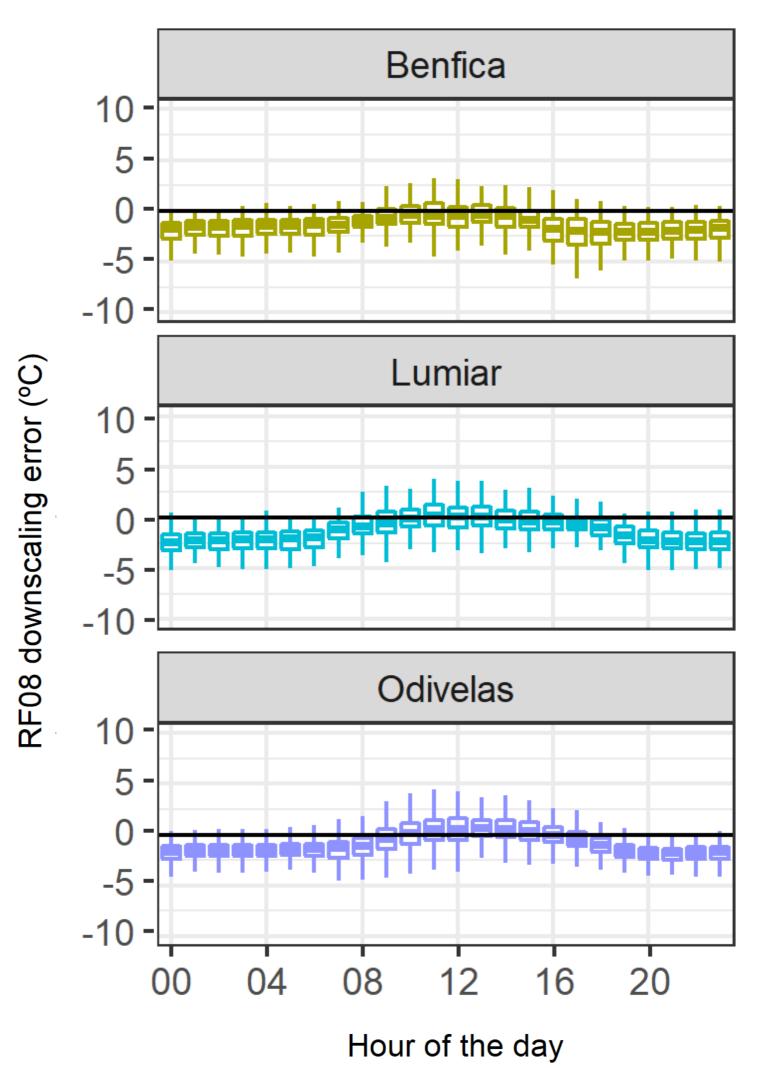
MAE (AROME) = 2.03



AROME forecast error (°C)

Hour of the day

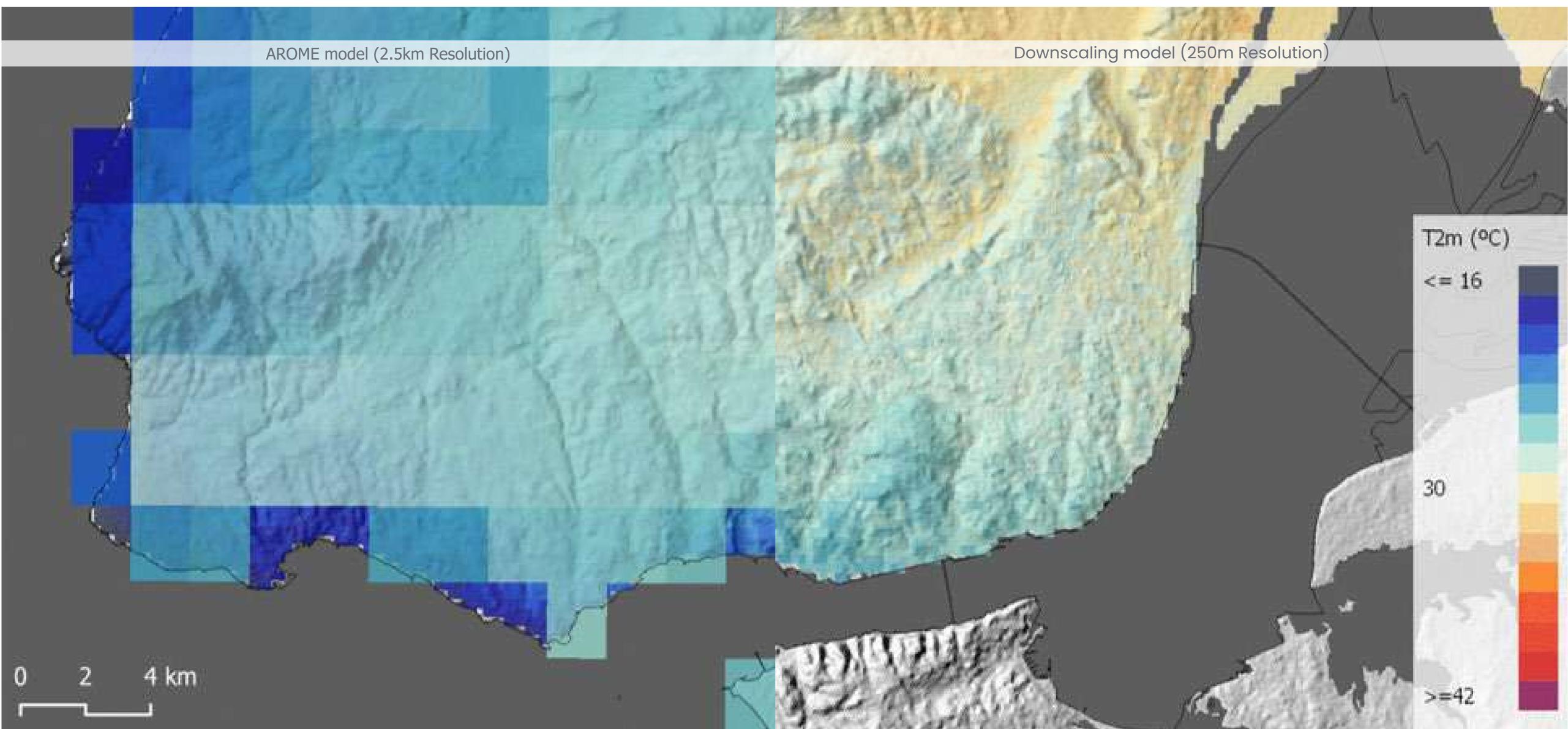
MAE (RF08) = 1.65

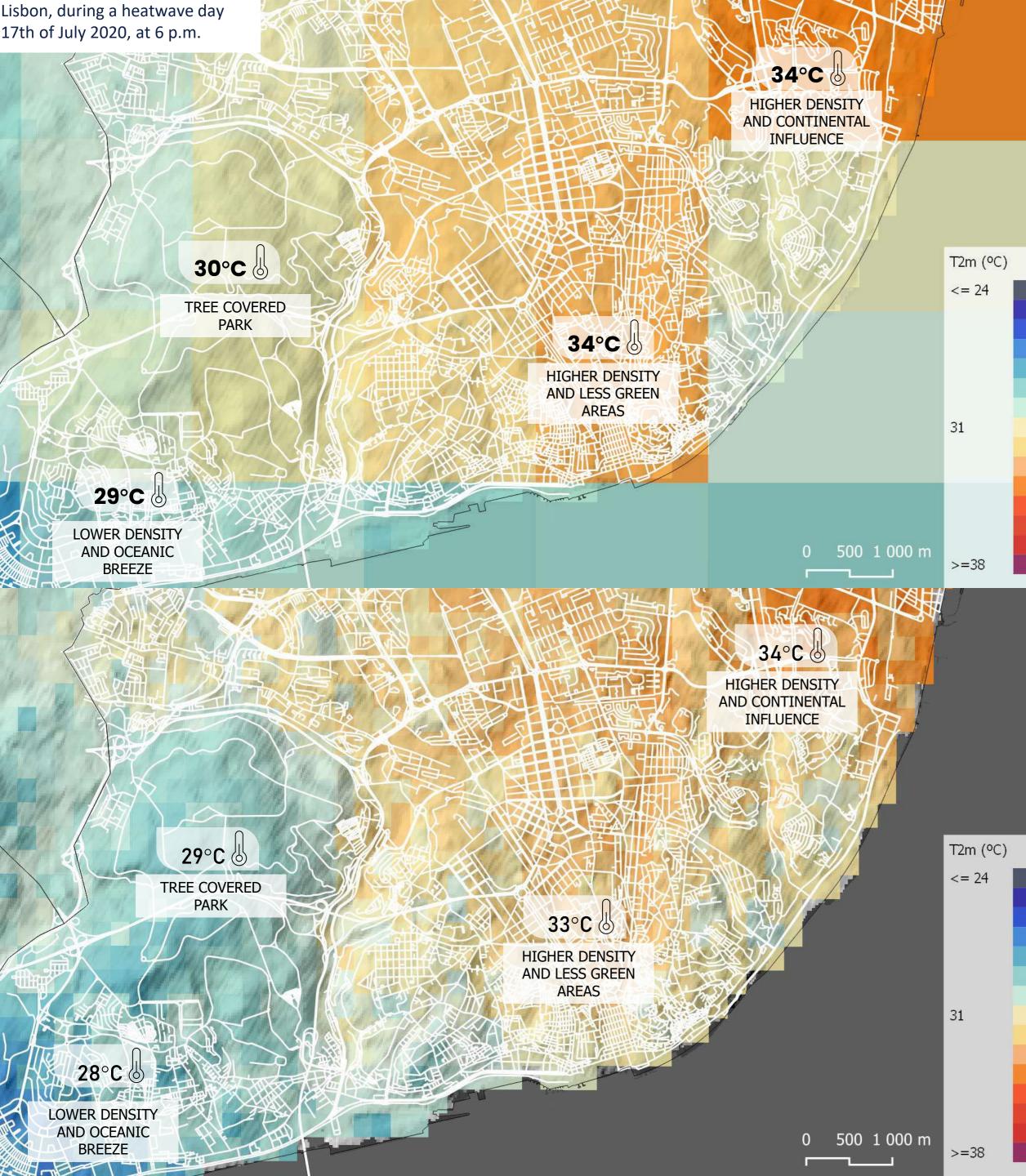




Results

Downscaling





Results

Downscaling

- Upsample AROME 10x (from 2.5 Km to 250 m)
- Maintain or Improve the AROME native error/bias

Netatmo stations: 20% improvement (0.4 °C)

WMO stations: Maintain **AROME'** s MAE (1.4°C); Slight

improvement in nocturnal hours and colder season (~0.2°C)

• Depict physically known thermal response (urban vs. green infrastructure)

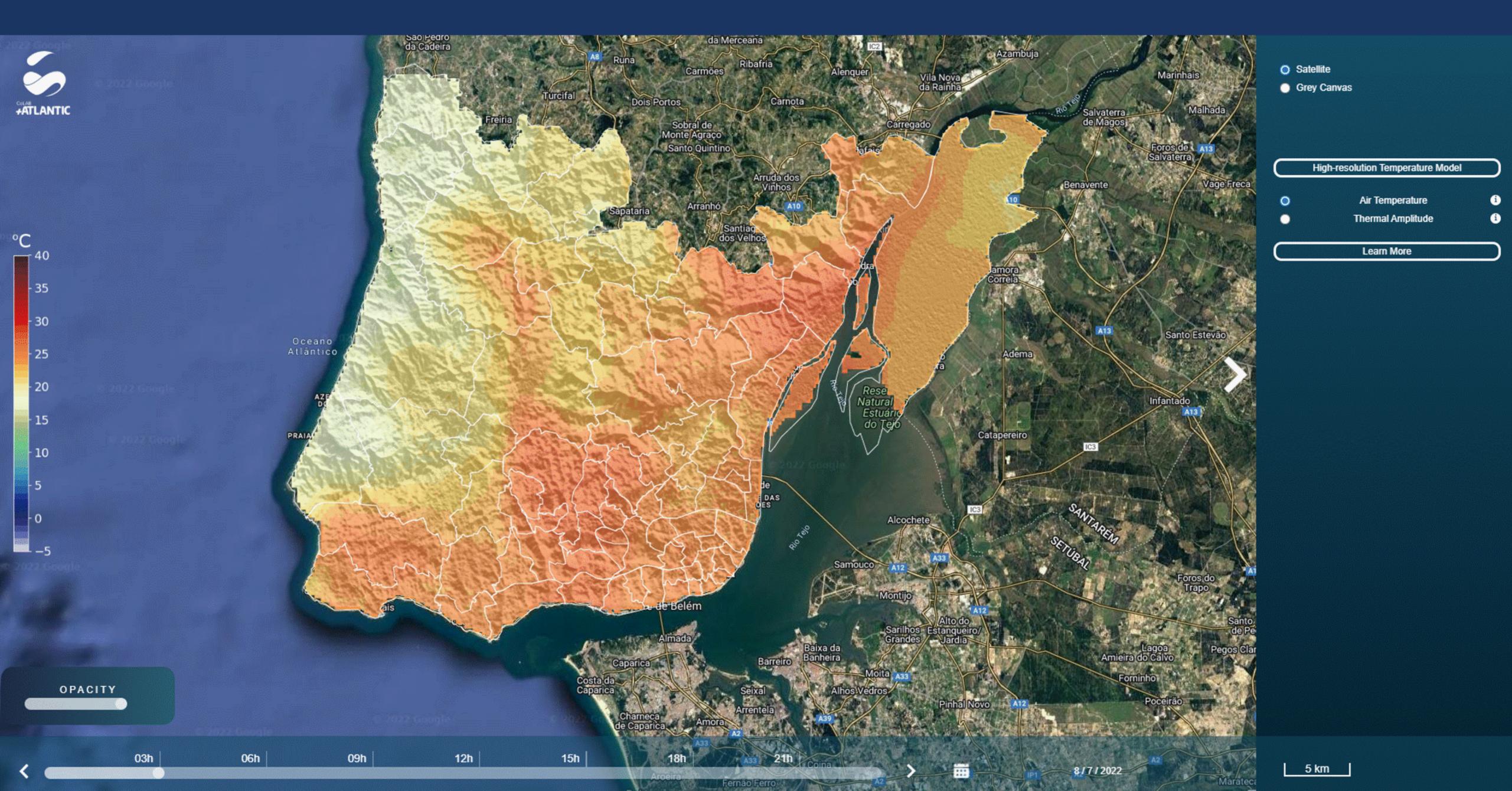


Mean 0.4°C improvement to AROME in nocturnal underestimation

Uphold or improve AROME deviations during heat/cold
extremes

Systematic improvement of at least ~0.2 °C from AROME in four HW/CW events in 2021

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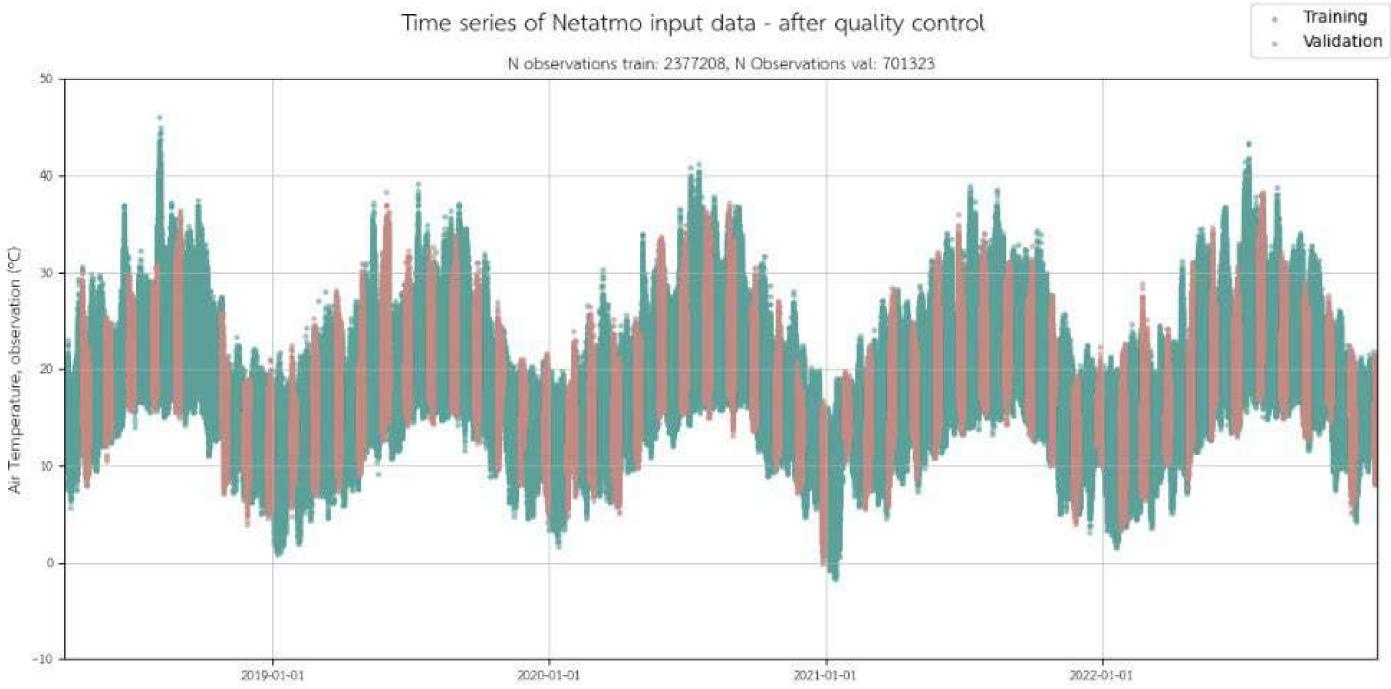


What Next?



Next Steps

Retraining Lisbon Model



More data (2018 – 2023)

Updated predictor layers

Data Augmentation for extreme cold/heat days

Predicting **AROME bias** instead of temperature

Preliminary work shows **improving accuracy**



Expanding the concept

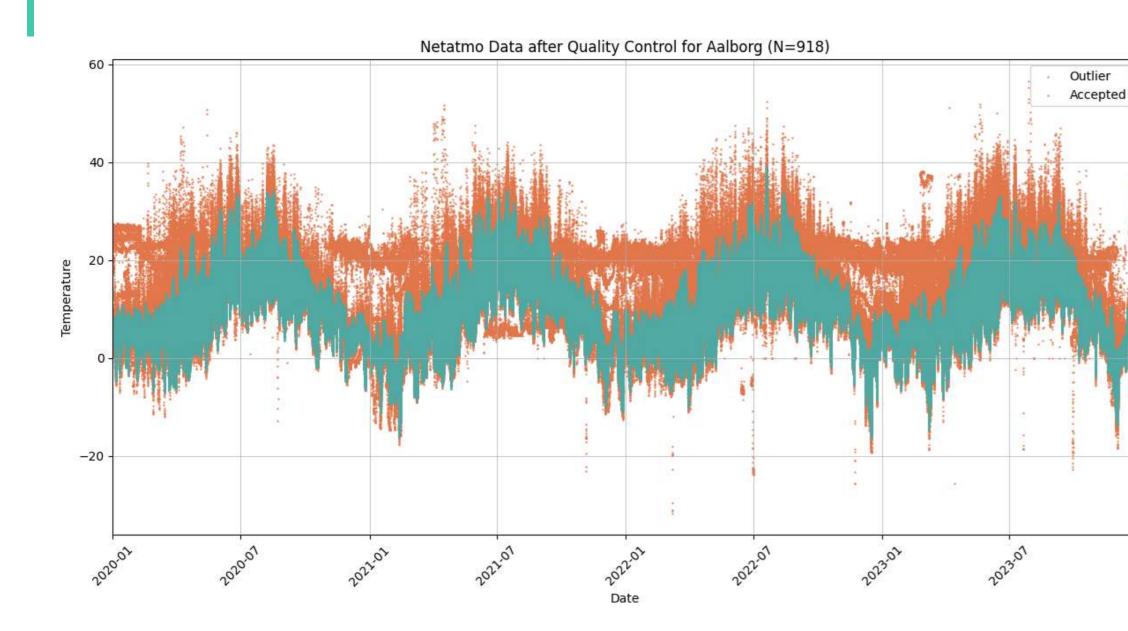
Expanding the proof of concept to **Denmark**

More **station density**

Test the **generalization in space**

Data Augmentation for extreme cold/heat days

Predicting **AROME bias** instead of temperature



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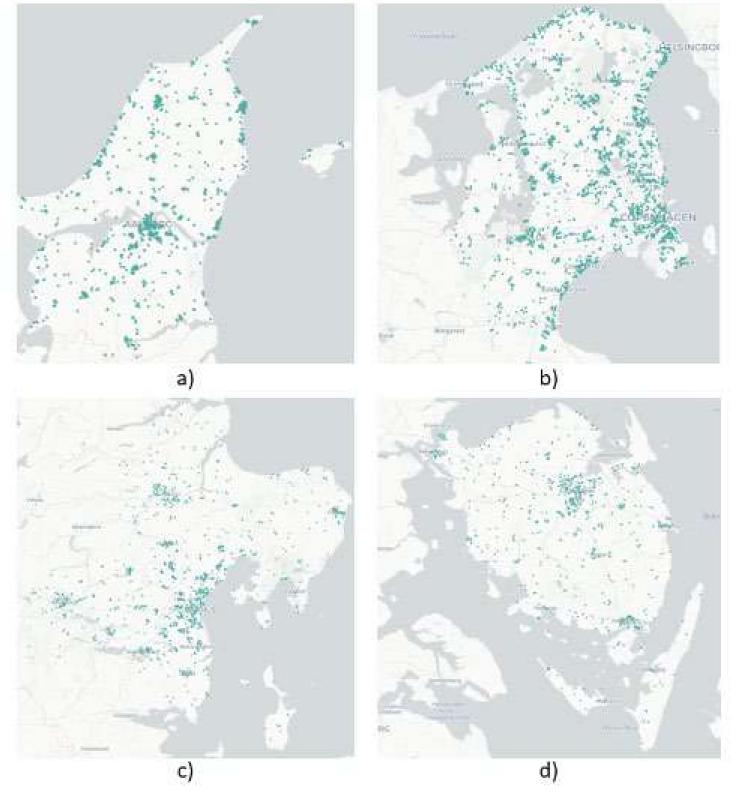
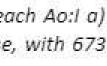


Figure 2. Maps with distribution of Netatmo stations with data within the four year period (2020-2023) for each Ao:I a) Aalborg, with 918 stations, b) Copenhagen, with 2223 stations, x) Aarhus, with 1220 stations and d) Odense, with 673 stations.



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